

Improving the Coastal Mean Dynamic Topography by Geodetic Combination of Tide Gauge and Satellite Altimetry

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Abstract

The ocean mean dynamic topography (MDT) represents the surface representation of the ocean circulation and may be determined by the ocean approach, which involves temporal averaging of numerical ocean circulation model information, or by the geodetic approach, wherein MDT is derived using the ellipsoidal height of the mean sea surface (MSS), or mean sea level (MSL), above the geoid. The ellipsoidal height of the MSS might be estimated either by satellite altimetry or at coastal tide gauges by connecting the tide gauge datums to the global reference frame by Global Navigation Satellite System (GNSS) positioning.

In this paper we present a novel approach to improve the coastal MDT where the solution is based on both satellite altimetry and tide gauge data using new set of 302 tide gauges with ellipsoidal heights through the SONE network.

The approach was evaluated for the Northeast Atlantic coast where a dense network of GNSS-surveyed tide gauges is available. With a “typical” misfit of around 9 cm, presumed to be mainly due to small scale geoid errors, single tide gauges place only weak constraints on the coastal dynamic topography, especially when the non-Gaussian nature of the errors means that much larger misfits are possible.

Keywords: mean dynamic topography, tide gauge, satellite altimetry

Introduction

The ocean’s mean dynamic topography (MDT) is the surface representation of ocean circulation. It is of interest to oceanographers to study the ocean’s surface currents (Wunsch and Stammer 1997), and to geodesists for e.g. linking height datums globally, particularly in regions where geodetic land-ties are not possible (Woodworth et al. 2013, Featherstone and Filmer 2012)

The ocean MDT may be determined by the oceanographic approach or the geodetic approach (Huang, 2017; Woodworth et al. 2013). The oceanographic approach involves the use of an ocean circulation models and the MDT is computed through temporal averaging over a given period.

The geodetic approach involves either satellite derived ellipsoidal mean sea surface models (MSS) (Andersen and Knudsen, 2009) or individual mean sea level (MSL)

observations from tide gauges expressed as ellipsoidal heights from co-located GNSS observations.

Figure 1 illustrates the way that the MSS or MSL is determined using geodetic methods like satellite altimetry and tide gauges. Once referenced to the same reference ellipsoid MSS and MSL represent the same “quantity”, but in the following we have used the terminology MSL for individual point observations (from tide gauges) and MSS for satellite derived grids.

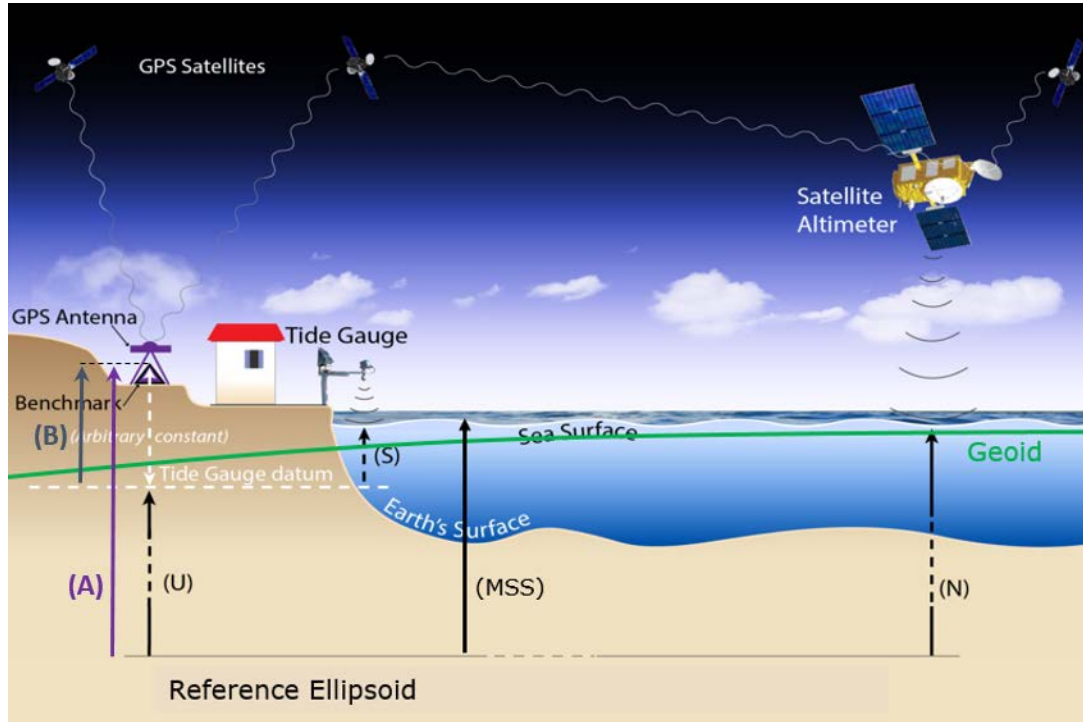


Figure 1: Satellite and tide gauge observations of the ellipsoidal Mean Sea Surface (MSS) or Mean sea level (S) relative to the tide gauge datum (U). The geoid height is called (N); (A) the GNSS-derived ellipsoidal height of a tide gauge benchmark in the global reference; (B) the benchmark with respect to the tide gauge datum. The figure is modified from Thierry Guyot (LIENSs).

From the ellipsoidal MSS or MSL data the MDT can be derived through a purely geometrical approach based on the simple equation

$$\text{MDT} = \text{MSS} - \text{N} \quad (1).$$

The MSS and/or MSL and geoid (N) must be given relative to same tide system and with respect to the same reference ellipsoid. We used the mean tide system and the TOPEX ellipsoid, with a semi major axis of 6,378,136.3 m and an inverse flattening of 298.257, but other ellipsoids or tide systems can be used through transformation. GNSS measurements, that are initially consistent with the WGS84 reference frame, are usually aligned with the International Terrestrial Reference Frame (ITRF) for Earth science

applications that require a high degree of accuracy (IERS, 2010). Thus, the GNSS ellipsoidal heights were here transformed to the most up-to-date ITRF realization at the time of the study (the ITRF08 (Altamimi et. al, 2011), which is associated with the GRS80 ellipsoid) and then converted into the mean-tide system following (Ekman, 1989)

This work continues the attempt to improve the coastal mean dynamic topography supporting ESA effort in the use of geoid information for height system unifications and connections (Gruber et al., 2012; Gerlach and Rummel, 2013; Amjardiparvar et al., 2013; Rummel 2012). The work is based on new geoid information from the ESA mission Gravity field and Ocean Circulation Explorer (GOCE) made available via the international center for global Earth Models (IGCEM, <http://icgem.gfz-potsdam.de/home>). We test a novel approach to derive an improved coastal MDT along the Northeast European coast using a geodetic combination of a satellite MDT at sea combined with a number of tide gauges derived MDT values along the coast. In the following, we refer to this combined geodetic approach as the GOCE++ approach as the work describes the result of the GOCE++ study initiated by the European Space Agency (ESA).

We chose a common five-year period 2003-2007 inclusive for deriving the MSS and MSL and consequently the derived MDT will represent this period. Initially, data and methods are described along with the way the MDT is determined geodetically from satellite and from tide gauges. We then present the updated set of 302 tide gauges with known ellipsoidal heights from the GNSS data assembly centre for the Global Sea Level Observing System at the [Système d'Observation du Niveau des Eaux Littorales \(SONEL, http://www.sonel.org\)](http://www.sonel.org). Here we evaluate the degree to which the point measurements of MDT at tide gauges can be reconciled with the broader scale MDTs of ocean models and those derived from altimetry.

Subsequently we present an attempt to merge satellite MDT and tide gauge MDT using an iterative method called GOCE++. The northeast European shelf has a huge number of tide gauges including 100 ellipsoidal tide gauges in the SONEL network. This enables us to perform a comparison of the effect on including tide gauges in the coastal MDT solution.

Currently, Synthetic Aperture Radar (SAR) altimetry is becoming available with Cryosat-2 and Sentinel-3 to improve the mapping of the short scales in the MSS. These data are an important step to improve future coastal MDT models and we present an evaluation of the MDT and improvements in the southern North Sea.

2 Data and methods

In the following sub-sections, we introduce the various datasets and methods to derive the MDT. We introduce the oceanographic MDT used for comparison as well as the satellite tide gauge data and the way geodetic MDT is computed.

2.1 Ocean MDT

Seven ocean models were available for computation of MDT in the study. These are three Nemo (Nucleus for European Modelling of the Ocean) ORCA (Madec, 2008) model integrations, one at a resolution of $1/4^\circ$ (NemoQ), and two at a resolution of $1/12^\circ$ with slightly different versions of the model code (Nemo12a, Nemo12b). Two Liverpool University implementations of the Massachusetts Institute of Technology (MIT) global ocean circulation model (Marshall et al., 1997a, 1997b), assimilating hydrographic information provided by the UK Met Office (Smith and Murphy, 2007); one in a coarse form (LivC), with a global resolution of 1° and a finer version (LivS) with an increased resolution of $1/5^\circ \times 1/6^\circ$ in the North Atlantic. The final two models are products of the ECCO consortium to calculate ocean state estimates by assimilating a wide variety of data, including geodetic data. EccoG (ECCO-Godae) has a global resolution of 1° (Köhl et al., 2007) producing a model state and evolution which is perfectly consistent with the model equations. Ecco2 (Menemenlis et al., 2005a; 2005b) is a finer resolution model (specified as approximately 18 km, but supplied on a 0.25 degree grid), with a looser assimilation scheme designed to match only certain patterns within the observations. All ocean models incorporate climatology for their initiation, as well as wind and atmospheric forcing from meteorological re-analyses. All models are averaged over the chosen period of 2003-2007 inclusive.

2.2 Satellite MDT.

Computing the satellite MDT is conceptually simple with the use of Eq.1. However, there are several complications in computing the satellite MDT. The major complication is the fact, that the geoid is typically given with limited resolution (typically hundreds of kilometers) compared to the MSS (tens of kilometers).

Subtracting a geoid N given in spherical harmonic coefficients up to a certain degree and order L the resulting dynamic topography estimate,

$$MDT_R = MDT + \Delta NL, \quad (2)$$

where the derived MDT_R consists of the MDT plus an un-modelled part of the geoid of spherical harmonic degrees and orders higher than L appearing as an omission error due to spherical harmonic coefficients which are omitted in the geoid model. This error decreases with increasing degree and order L of the geoid. The GOCE satellite has been paramount to reduce the geoid error in global geoids where the degree and order of satellite only GOCE geoids is typically around 250.

Consequently, a filter needs to be derived and applied to Eq. 2 to eliminate the ΔNL without removing real MDT signal in the filtering. It is naturally important to design appropriate linear or non-linear filters for this operation (Bingham and Haines, 2006; Bingham et al., 2011; Sanchez-Reales et al, 2016).

We decided to use the alternative approach to limit the omission error ΔNL by using a geoid and a MSS given at the same resolution. We used the European Improved Gravity model of the Earth by New techniques (EIGEN) combined 6C4 geoid to degree and order 2190 due to consistency with the MSS. This geoid consists of a combination of GRACE and GOCE up to degree and order 260 augmented with the DTU10 surface gravity data. For the investigation, we used the DTU10MSS (Andersen and Knudsen, 2009; Andersen et al., 2016). This is computed as an average of satellite altimetry over the period 1993-2009, but is mapped to the 2003-2007 average by using the difference in AVISO (Archiving, Validation, and Interpretation of Satellite Oceanographic data) absolute dynamic topography (ADT) averages over the two periods. DTU10MSS was preferred because this model is consistent with various geoids like the EIGEN6C4 and the TUM13 (Fecher et al., 2015) geoid (the gravity anomaly dataset is derived from the DTU10MSS, so the difference does not add artificial small-scale errors into the dynamic topography).

Altimetric MSS is corrected for the ocean's inverted barometer (IB) response (static atmospheric loading effect) using Wunsch and Stammer (1997) and a correction for periods shorter than the sampling of the Topex and Jason satellite via the Dynamic Atmosphere Correction. This correction is applied, as it reduces the sea surface variability leading to a more stable estimation of the mean. In our approach, we decided to restore the effect of the atmosphere on the sea surface to be consistent with the tide gauge mean.

The total error of the satellite derived MDT will be a combination of the geoid omission errors along with errors in both the MSS and the geoid coefficients (commission errors). The satellite MDT is typically less accurate in the coastal zone due to a combination of less accurate altimetric observations, less accurate geoid mapping due to large bathymetry/topography variations and the fact that the MDT changes more rapidly in the coastal zone.

2.3 Tide gauge MDT

Using global tide gauges for comparisons and combination with satellite MDT implies that the tide gauges height is linked into a global reference using GNSS measurements at the local tide gauges transforming the observations to ellipsoidal heights. This link usually relies on two quantities: the GNSS-derived ellipsoidal height of a tide gauge benchmark in the global reference (A in Figure 1), obtained from the processing of permanent or episodic GNSS measurements, and the height of this benchmark with respect to the tide gauge datum (B in Figure 1). The latter is always known since the full tide gauge benchmark datum history is required for the tide gauge records of the RLR databank from the Permanent Service for Mean Sea Level (PSMSL) (Holgate et al, 2013), from which the MSL values from tide gauges were retrieved in this study. When the former is not available, then a supplemental quantity is needed to connect this benchmark to one where GNSS ellipsoidal height is known (mostly using spirit leveling). This is commonly called a geodetic tie (Woodworth et. al, 2017), and ultimately yields A in Figure 1.

2.3.1 Tide gauges with permanent GNSS.

The use of GNSS for monitoring the tide gauge sites stability is optimized when it is processed continuously and globally (Wöppelmann and Marcos, 2016). Thus, the

availability of positions and vertical velocity of tide gauge co-located permanent GNSS stations was priority investigating. The geodetic ties of 113 RLR tide gauges with nearby permanent GNSS stations were collected from the SONEL data assembly centre (<http://www.sonel.org>). Fourteen additional ties were recovered for the German tide gauges, which are not in the RLR dataset yet (<http://pegelonline.wsv.de>). For most of these permanent GNSS stations, the ellipsoidal height (associated to an epoch and an uncertainty) was extracted from the last GPS solution (ULR6a) of the ULR analysis centre provided on SONEL (Santamaria et. al, 2017). The vertical velocities were the used to propagate the ellipsoidal height from the reference epoch to the average epoch of the selected period 2003-2007 (2005.5). For permanent stations, that were not included in this solution, the ellipsoidal heights were obtained from the average of three positions per day, processed using the Canadian CSRS-PPP tool (<https://webapp.geod.nrcan.gc.ca/geod/tools-outils/ppp.php>), also expressed in the ITRF08 reference frame. Similarly as for the ULR6 stations that did not satisfy the “robust” velocity criteria, the propagation to the epoch 2005.5 could not be performed for these stations.

2.3.2 Tide Gauges with episodic measurements.

Ellipsoidal heights from GNSS campaigns at tide gauge sites were retrieved from different sources, mostly via Woodworth et. al (2013), Woodworth et. al (2015), Featherstone & Filmer (2012), Lin et al (2015) and SONEL, and led to 189 additional RLR ellipsoidal heights.

For most of the ellipsoidal heights coming from episodic GNSS measurements, the epoch and the uncertainty is unknown. This is a source of error, but there was no means to propagate the heights to the mean epochs of the selected period (they were adopted as is, as for permanent GNSS stations without enough data to obtain a robust velocity and the heights coming from the CSRS-PPP tool), and a 3 cm uncertainty was associated by default (arbitrary based on our personal experience).

2.3.3 Computing the mean sea level

For the RLR stations, the monthly time series from PSMSL were used to compute the MSL over (2003-2007). For records with more than 70% of data over this period, a classic average was calculated. For the remainder 54 stations we applied simultaneous

observations from satellite altimetry data to fill the gap. This was done in three stages: first, a least squares fit of IB-corrected tide gauge data on an annual cycle, semiannual cycle, linear trend, and altimeter time series was computed for every altimeter point (using the AVISO gridded dynamic topography, www.aviso.altimetry.fr/en) within 150 km of the tide gauge. Second, the altimeter point for which the resulting fit explained the largest percentage variance of the tide gauge data was selected, and tide gauge data predicted from that least squares fit for every month. Third, that predicted tide gauge time series was used to fill gaps in the tide gauge data, with an additional linear trend added over each gap to ensure a perfect match at the end points. This yield a dataset of 302 ellipsoidal MSL values for the 2003-2007 period globally distributed (Figure 2).

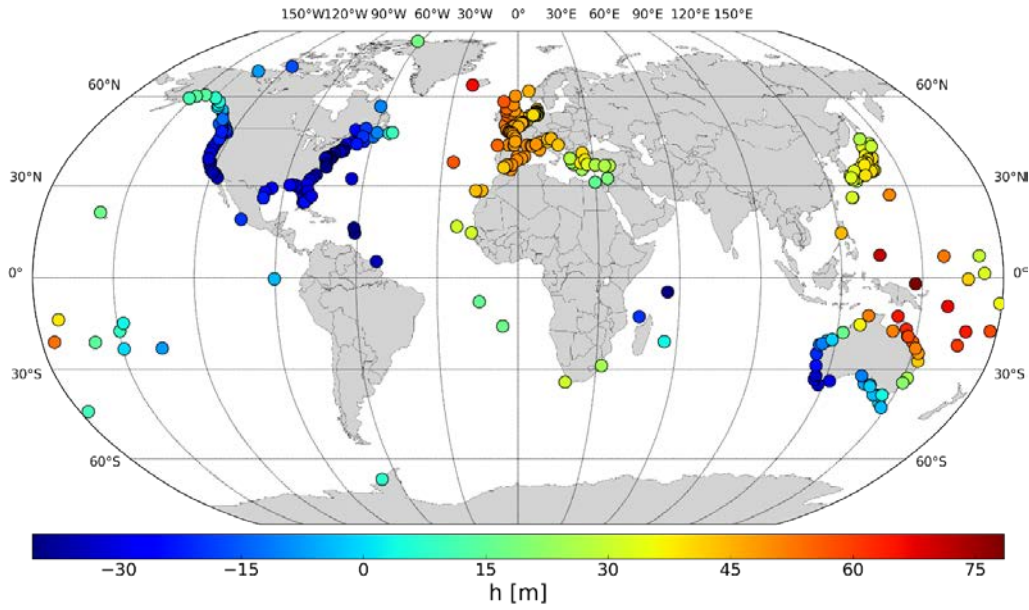


Figure 2: Ellipsoidal Mean sea level at tide gauges over (2003-2007)

Following Equation 1, the MDT computation at a tide gauge implies subtracting the geoid height from the ellipsoidal MSL. Thus, at a tide gauge site, Equation 1 can be written as follows:

$$MDT_{TG} = EMSL_{TG} - GEOID_{TG} \quad (3)$$

where MDT_{TG} is the MDT value at the tide gauge, $E_{MSL_{TG}}$ is the tide gauge ellipsoidal MSL and $GEOID_{TG}$ is the geoid value at the tide gauge. $E_{MSL_{TG}}$ can be developed as follows:

$$E_{MSL_{TG}} = E_{datum} + MSL \quad (4)$$

where E_{datum} is the ellipsoidal height of the datum of the tide gauge measurements and MSL value as observed by the tide gauge with respect to this datum.

E_{datum} is obtained by adding the ellipsoidal height of the nearby GNSS station (E_{GNSS}) and the ellipsoidal height difference between both ($DH_{dat-gnss}$):

$$E_{datum} = E_{GNSS} + DH_{dat-gnss} \quad (5)$$

$DH_{dat-gnss}$ is equal to the geodetic tie used ($tie_{dat-gnss}$) only if it comes from differential GNSS (and the tie is geometric and expresses indeed a difference of ellipsoidal heights). If the geodetic tie comes from spirit leveling, it expresses a difference of elevations and the geoid gradient has to be taken into account:

$$DH_{dat-gnss} = tie_{dat-gnss} + GEOID_{TG} - GEOID_{GNSS} \quad (6)$$

where $tie_{dat-gnss}$ is the geodetic tie used coming here from spirit leveling and $GEOID_{GNSS}$ is the geoid value at the tide gauge-co-located GNSS point from which the tie has been derived. At the end, Equation (3) becomes:

$$MDT_{TG} = E_{GNSS} + tie_{dat-gnss} + MSL - GEOID_{TG} \quad (7)$$

if the $tie_{dat-gnss}$ is a difference of ellipsoidal heights, and:

$$MDT_{TG} = E_{GNSS} + tie_{dat-gnss} + MSL - GEOID_{GNSS} \quad (8)$$

if $tie_{dat-gnss}$ is a difference of elevations.

Therefore, computing the MDT values at the tide gauges from the ellipsoidal MSL implies identifying the type of the geodetic tie in order to determine whether the geoid value has to be extracted at the tide gauge position or at the co-located GNSS point. Concerning the ellipsoidal height coming from episodic GNSS, it was assumed that the GNSS point was situated in the vicinity of the tide gauge, which induces a negligible geoid slope between both instruments. Thus, for tide gauges with permanent GNSS stations the geodetic ties were explicitly accounted for, but for episodic GNSS it was assumed (for lack of any other information) that no tie was necessary. In strong geoid gradients areas, the impact could reach more than ten centimeters (Figure 13). The importance of the geoid extraction point indicates the equal importance of the correct position of the tide gauge: particular care has been taken to check the coordinates of the 302 selected tide gauges.

3. Evaluation and representativeness of tide gauge MDT

The tide gauge measurements with geocentric position information discussed above have been converted to MDT estimates by subtracting a geoid from the MSS thus defined, i.e. by applying equation (7) or (8) as appropriate. Here we evaluate the degree to which the point measurements of MDT at tide gauges can be reconciled with the

broader scale MDTs of ocean models and those derived from altimetry, considering the effect of different geoids and different truncations. Note that, for this comparison with models (most of which do not include atmospheric pressure as a forcing), we again apply the IB correction to all MDTs derived from tide gauges or altimetry.

As the global mean value of MDT from ocean models is ill-defined, we subtract this off from all mapped products. To do this in a consistent way given the different spatial domains of the different products, we first choose a reference model with global coverage (Ecco2) and subtract off the global mean MDT from this model. For other mapped products, we subtract off the spatial average of the difference from the de-meaned Ecco2, over their common domain. For comparisons purely at tide gauge positions below, we further subtract the median of differences from the tide gauge MDT across all available sites, so statistics relate to the spatial variations in MDT and not the absolute values.

In addition to the Aviso MDT, we also use a second altimeter-derived MDT labelled here TUM13, which is formed from the DTU2010 MSS minus the TUM13 geoid. This was smoothed with a spatially-varying Wiener filter chosen using a signal size estimate from the Nemo12a ocean model and noise based on the assumption that variations in a known smooth region of the Pacific are all noise. The same product was also used in Filmer et al. (2018).

The ordering of the tide gauges follows the PSMSL coastal ordering. This starts with Norway, running anticlockwise around Europe, Africa and Asia, then covers Australia and the Pacific, before running anticlockwise around the Americas starting with Alaska and ending with Arctic Canada, before finally moving to Antarctica (the last point only). See Appendix 1 for more detail.

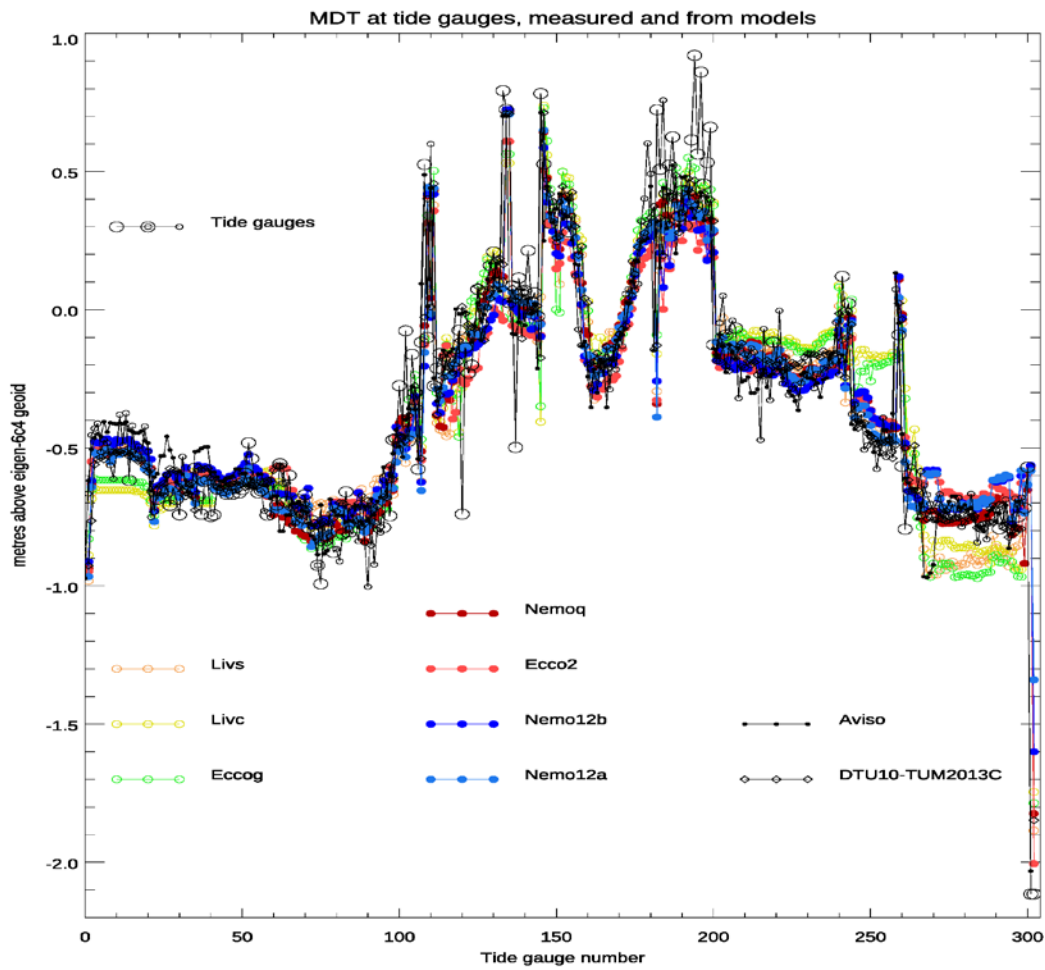


Figure 3: The MDT at 302 tide gauges (black open circles) using the EIGEN-6C4 geoid to its full resolution, compared with predictions from 7 ocean models (colours), and 2 satellite-derived MDTs (black). Large open circles represent sites with continuous GNSS, and small open circles those with episodic GNSS. Continuous GNSS, but an estimated MSL/MSS error of > 0.04 m is shown as a double circle (5 sites).

Several considerations can be made immediately from Figure 3. The match is generally good, but the tide gauges show a systematic high bias between about numbers 182 and 200, corresponding to the Pacific Islands. The two prominent downward spikes (121 and 138) are Aburatsubo (near Tokyo) and Mikuni (fairly nearby, but on the opposite, north coast of mainland Japan). Tectonic activity is an obvious concern, but the many other Japanese records look good. Other “spikes” are common to tide gauges and models, and represent excursions off the main coastline to islands.

Table 1 shows summary statistics based on these comparisons. The distribution is clearly non-Gaussian, with long tails and a more compact central region, leading to a high excess kurtosis (kurtosis-3; a Gaussian distribution has an excess kurtosis of zero, <https://www.itl.nist.gov/div898/handbook/eda/section3/eda35b.htm>). We see that the global MDTs which produce the worst comparisons tend to have lower kurtosis, becoming closer to Gaussian values. This provokes a tentative interpretation, that the global MDT errors tend to be more Gaussian, and the errors at tide gauges include more extreme values, probably due to missing fine-scale geoid information. This will be investigated in more detail below.

Table 1: Statistics comparing the coastal MDT at tide gauges estimated using the EIGEN-6C4 geoid to full resolution, against 8 different global MDTs extrapolated to the same positions. Height values are in metres. The pdfs are non-Gaussian, so they are characterized by half the range which contains the number of values which would be expected to fall within 1 sigma in a Gaussian distribution (1 sigma equiv), or quarter the range for 2 sigma. Also given are the minimum and maximum values (TG-MDT after subtracting the median), and the number of missing values (because some MDTs do not have data in some regions). Where there are no missing values, there are 302 points being compared.

Global MDT	Std	1 sigma	2 sigma	Min	Max	Skewness	Excess Kurtosis	N missing	% err < 9 cm
Nemo12a	0.132	0.097	0.142	-0.775	0.571	-0.33	7.69	0	67
NemoQ	0.123	0.095	0.127	-0.588	0.617	+0.43	5.41	0	68
Aviso	0.134	0.099	0.131	-0.727	0.556	-0.21	6.27	3	65
DTU10TUM13	0.131	0.109	0.139	-0.726	0.589	+0.10	4.47	4	62
Nemo12b	0.139	0.101	0.157	-0.586	0.633	+0.60	4.45	0	64
Ecco2	0.145	0.121	0.153	-0.471	0.706	+0.99	3.18	1	54
Livs	0.161	0.132	0.177	-0.632	0.753	+0.75	3.24	3	54
EccoG	0.158	0.150	0.159	-0.600	0.540	-0.14	0.63	7	40
Livc	0.163	0.153	0.169	-0.572	0.546	-0.05	0.62	7	42

It is clear, that the data are able to discriminate between models, especially when ignoring the tails of the distributions. By far the most important factor is the quality of the geoid, and in particular its fine scale structure corresponding to limiting the omission error in Eq 2. This is similar to what was seen by Huang (2017). This is illustrated by Figure 4, which repeats Figure 3 but with the EIGEN6-C4 geoid truncated

at degree 300 roughly corresponding to the degree and order of a satellite only geoid from GOCE.

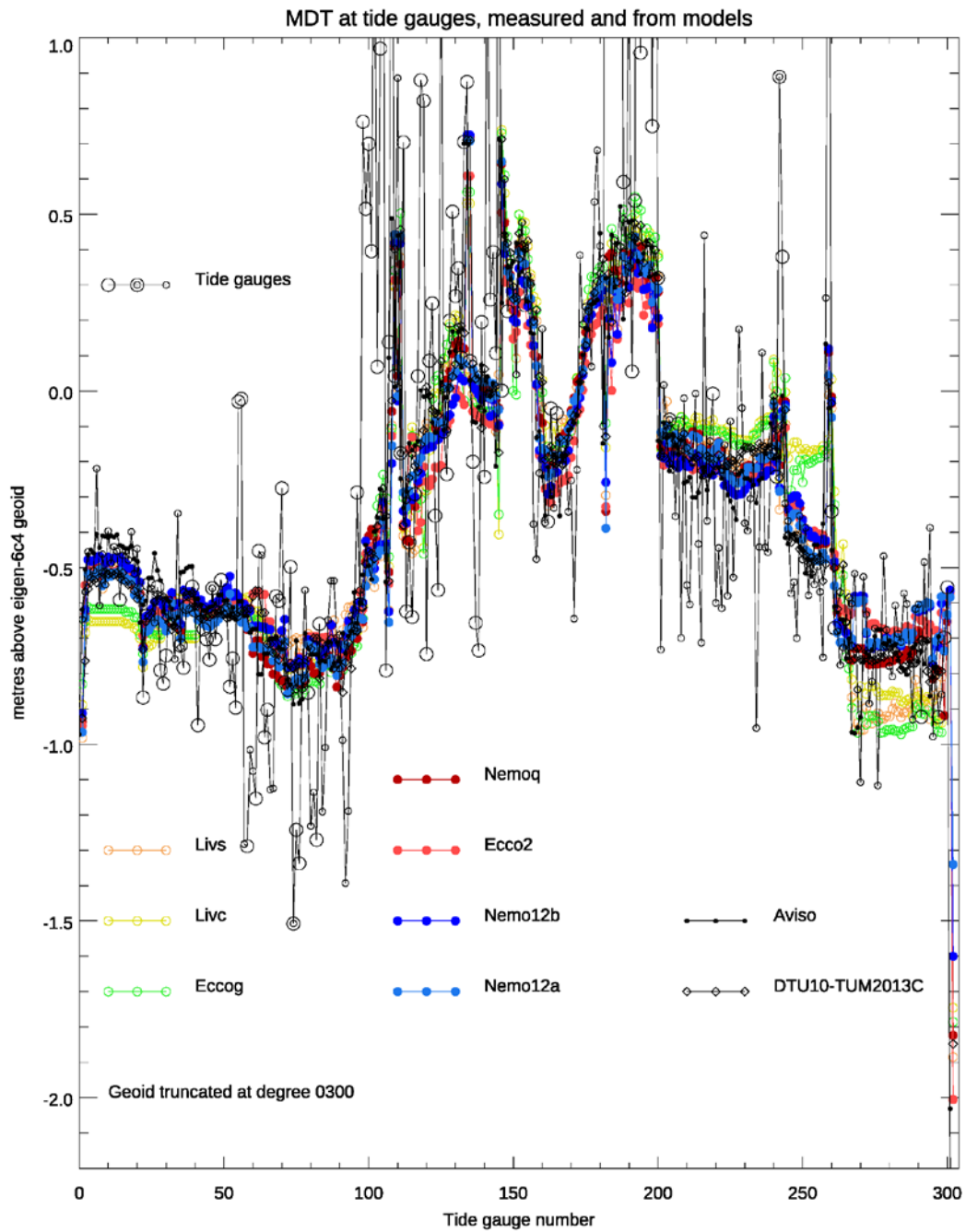


Figure 4: repeat of Figure 3 but with the EIGEN-6c4 geoid truncated at degree 300.

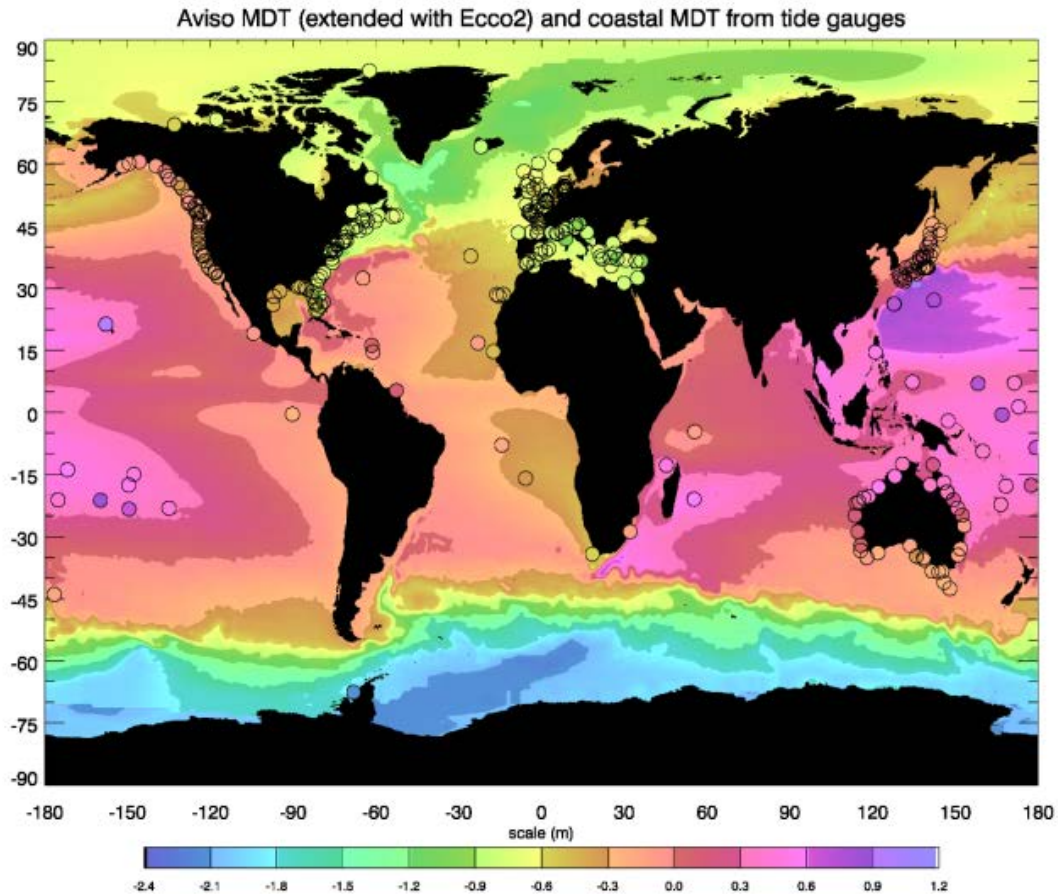


Figure 5: The spatial pattern of the Aviso MDT compared with the tide gauge MDT based on the EIGEN-6c4 geoid. Where the Aviso MDT has no data (mainly ice-covered regions), the Ecco2 model is shown instead.

The spatial distribution of the various MDT measurements can be seen from Figure 5. The map cannot show all details, however the general patterns are visible. We note that the MDT at the gauge on the east coast of South Africa (Richards Bay) agrees well with the Aviso global MDT, however the resolution of the thin strip of ocean to the west of the Agulhas current is crucial, as nearby ocean values are very different. More detailed examinations reveal similar results for the Gulf Stream, Kuroshio, and East Australian Current.

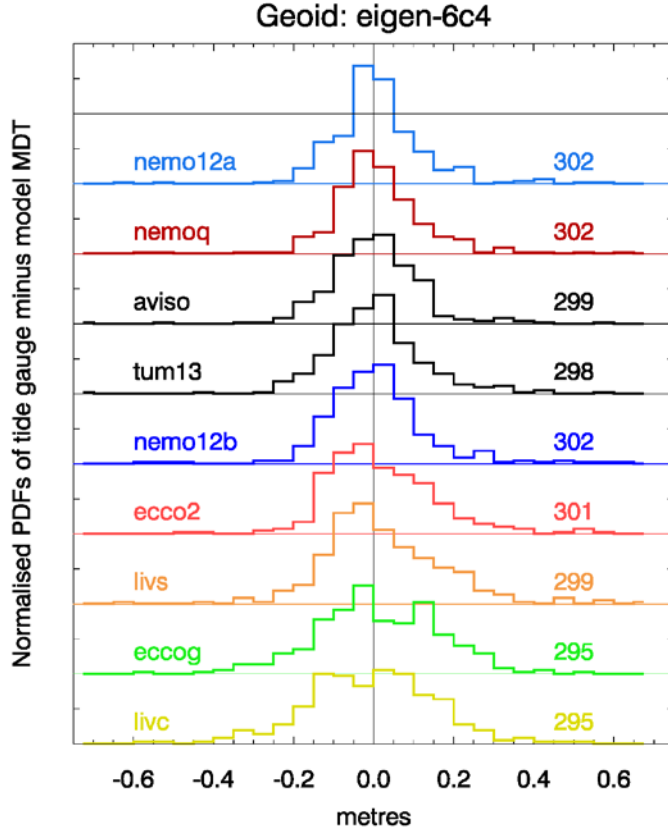


Figure 6: PDFs of the misfits between the tide gauge MDT using the EIGEN-6C4 geoid to its full resolution, and the various global products (median values subtracted). The PDFs are normalised relative to a Gaussian with the same standard deviation, and are offset by ± 1 from their neighbours. Numbers to the right represent the number of available points for comparison in each case.

The distribution of misfits for these tide gauge values relative to the various global oceanographic MDT models is shown in Figure 6. A broader, more Gaussian distribution (peaks close to 1 with this normalisation) are found with the coarse resolution models (livc,livs,eccog). It also illustrates the fact that the better comparisons (narrower distributions) have higher peaks than would be expected for a Gaussian distribution, consistent with the high kurtosis discussed above.

The PDFs in Figure 6 appear to have a longer positive tail than the associated negative tail. A long positive tail suggests that tide gauges are at positions, which are special in some way, leading to a systematic sign in the misfits, where they are large. We hypothesize that this is reflecting the fact that, being coastal, many tide gauges are close to the top of a steep continental slope, which produces a local geoid maximum at small length scales. In the absence of sufficient in situ gravity data, such maxima are likely to

be reduced by smoothing, leading to the geoid product estimate being lower than the true geoid, and hence the MDT being higher than the true value.

With this interpretation, high kurtosis and a longer positive tail in the PDF would be diagnostic of an artificially smooth geoid. We can test this interpretation by calculating the statistics using various degrees of smoothing, produced by truncating the geoid at different resolutions. At the same time, this allows us to see the effect of resolution on the size of the misfit. The results are shown in Fig 7 showing that a long positive tail and high kurtosis do occur when the geoid is poorly resolved, supporting our interpretation. The effect seems to peak at truncations of around degree 300-600, corresponding to length scales of about 40-70 km, but remains even with the full resolution geoid. We interpret this as being a measure of the missing in situ information limiting the resolution of the geoid in some places.

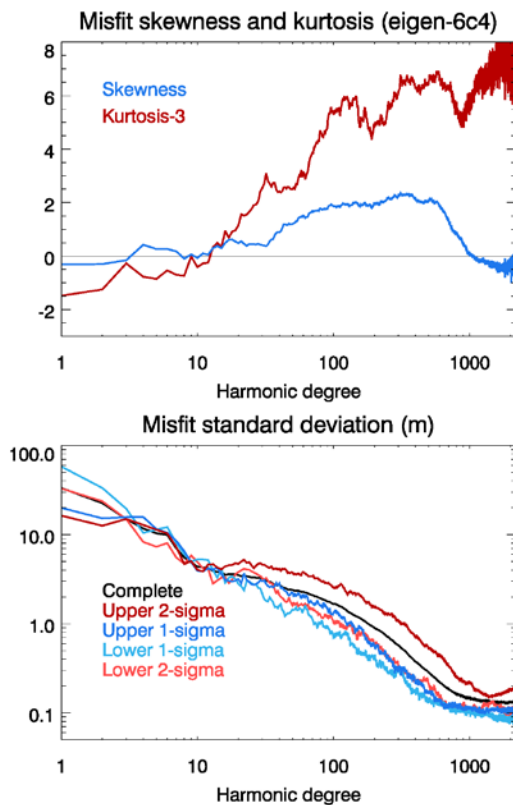


Figure 7: Statistics of misfit between Nemo12a and tide gauge MDTs using the EIGEN-6c4 geoid at different truncations. The standard deviation is shown based on the full PDF, or as the width of the PDF required to contain 34% of data below or above the median, or half the width containing 47.5% below or above the median. These would all be the same for a Gaussian distribution.

The error standard deviation results show large and continuous improvements with increasing resolution, but begin to plateau at around degree 700, again suggestive of a limit to the added value of the in situ data. The values taken from different parts of the distribution are quite consistent (and lower than the overall standard deviation), apart from the value representative of the upper, more extreme values, confirming that the positive tail has the larger departure from a Gaussian distribution.

In an attempt to determine which are the best models and geoids to use, we rank the 27 possible comparisons of 9 global MDTs with 3 geoids, according to the size of the misfits. The global MDTs include the 7 ocean models described in the Methods section, plus two observational products: Aviso and TUM13.

The non-Gaussian nature of the PDFs means that no single statistic is representative of the distribution, so we use three different measures: standard deviation, width of the part of the PDF containing the central 68% of the data, and width of the part containing the central 95%. The resultant rankings are shown in 8. The three geoids are EIGEN-6C4, GOCO05c and EGM2008. The GOCO05c product was extended from its native resolution of degree 720 using EGM2008 coefficients at higher degrees (Pavlis et al., 2012). All three geoids therefore include information up to degree 2190.

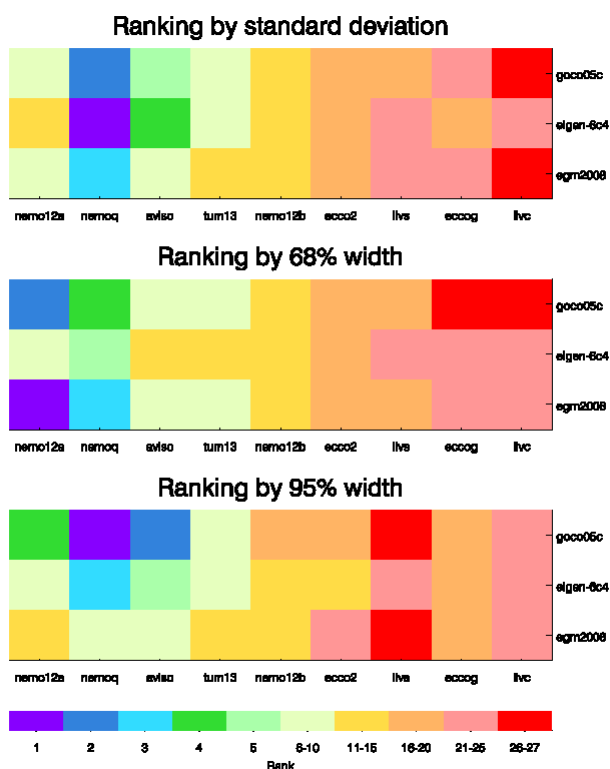


Figure 8: Ranking of quality of comparison of MDTs at tide gauges using three geoids, compared with 9 global products. Three different criteria are used as described in the text

The low-resolution ocean MDT models, LivC, EccoG and LivS, consistently perform poorly, and NemoQ, Aviso and Nemo12a consistently perform well, with TUM13 and Nemo12b in the middle, and Ecco2 closer to the poor end. It is perhaps to be expected that the “observational” products, Aviso and TUM13, should do well, being the most strongly constrained to the real ocean. It is also natural that TUM13 should be slightly worse, as it does not include any dynamical constraints, unlike AVISO. However, the ocean model results are interesting. The three low-resolution models perform almost equally poorly, despite assimilation of geodetic data into one of them (EccoG).

The ranking of geoids is less consistent (unsurprisingly given the importance of fine scale information which is derived from very similar, incomplete sources in all cases), with different geoids performing best under different ranking criteria.

Looking back at Table 1, the final column shows the percentage of comparisons which were within 9 cm, after matching the medians. For a Gaussian distribution, this would be 68.2% if 9 cm represented one standard deviation. We can see from this that the best global MDTs have “typical” errors of around 9 cm, if “typical” is defined as analogous

to one standard deviation of a Gaussian distribution. This error is similar to the findings for Norway as presented by (Ophaug et al., 2015)

4. GOCE++ combined Coastal MDT

The satellite derived geodetic MDT is generally less accurate in the coastal region due to the sparse coverage of valid altimetry observations. Artifacts in the coastal region are typically seen as MDT contours crossing the coastline when plotting these.

In the GOCE++ project we attempted to derive a coastal MDT by inserting MDT values over land and subsequently smoothing these into the ocean using an iterative spatial filter. The land values can be based on the tide gauge MDT values, the satellite based MDT in the coastal grid cells, or a combination of both.

4.1 Land values

In the tide gauge based solution the tide gauge MDT values are used to fill in the land values. This is done by linearly interpolating the individual MDT values onto the entire grid of land cells.

In the satellite based solution the land value for a given coastal grid point is estimated as the average of the MDT grid values in a box centered on the coastal grid cell. The size of this box is by default chosen to be 9 times 9 grid points which approximately corresponds to a 1 times 1 degree box, as all grids for the test is at 1/8 degree resolution. Hence a local average (defined by the size of the box) for each coastal grid cell is used as a representation of the land value. It is important to notice that the altimetry based land values will not have the constraining effect provided by the tide gauge MDT values, but solely have an additional smoothing effect ensuring a less noisy MDT along the coast.

In the combined solution, coastlines without tide gauges are supplemented with altimetry based land values as described above. Subsequently the land values along the coast are simply interpolated to all land grid cells using linear or nearest neighbor interpolation.

4.2 Filtering

To smooth the raw MDT values a spatial filter is applied. The filter is an average or box filter where the kernel is a n_x times n_y matrix, where n_x and n_y are the number of

kernel points in the east-west and north-south direction, respectively. In the north-south direction, the size of the filter is fixed with a default value of two while it is scaled by the latitude in the east-west direction. The filter is iteratively applied over both land and ocean grid cells causing the simple average filter to converge toward a Gaussian filter. However, at each iteration, the land values are reset to their original value. This has the effect of maintaining the coastal values and increasing the smoothing in the coastal regions without affecting the open ocean. In general ten iterations are used.

4.3 Software

The methodology was implemented in the publically available software “coastMDT” which allows estimation of the coastal MDT in a given region (see the detailed description of the software and how to access it in Appendix 2). Figure 9 shows a flowchart of the software “coastMDT”. The green boxes represent data input to the software; the purple boxes the functionality and the blue box the final product. The functionality can briefly be explained in the following steps:

- Step 0: The tide gauge data is references to the same ellipsoid as the altimetry data.
- Step 1: The raw altimetry and tide gauge MDTs are derived.
- Step 2: A region of interest is selected based on a longitude and latitude range.
- Step 3: The land values are estimated based on the methods described above.
- Step 4: Filtering over both land and ocean grid cells is applied.
- Step 5: Plotting, error estimation, and saving of the final MDT.

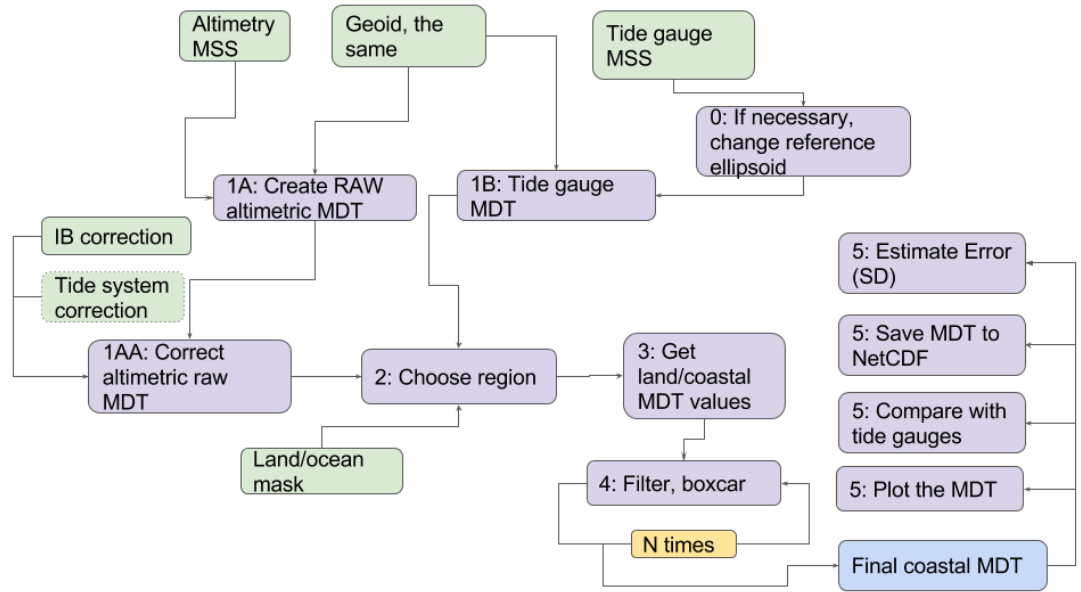


Figure 9: Flowchart of the software “coastMDT”.

5. Results

5.1 The GOCE++ Northeast Atlantic MDT model

As a demonstration of the combined coastal MDT based on the GOCE++ approach the raw and filtered solutions are shown for the Northeast Atlantic coastline (Figure 10). The coastal values in the example are based on a combination of the tide gauge and altimetry as described above. The figure also show the grid number labelling of the coastal tide gauges along the NW European coast (excluding Britain) used in the following evaluation of the MDT. The tide gauges used in this section all lie in the range of numbers 2 to 98 as used in Figures 3 and 4 and described in appendix 1.

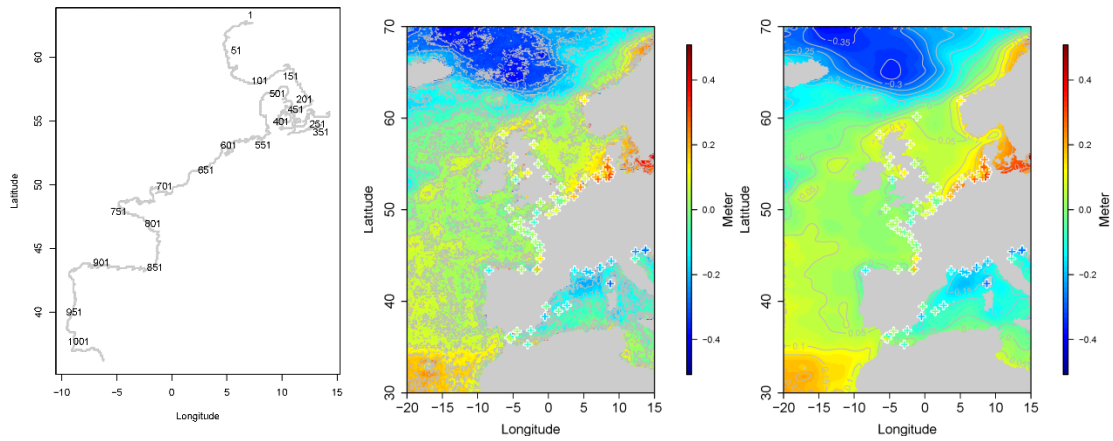


Figure 10: Northeast Atlantic tide gauges and MDT, Left the numbering of the grid points used for the evaluation of the GOCE++ model. Middle. An example of the raw MDT, right, the filtered MDT, where the land values are based on a combination of tide gauges and altimetry. The location of the tide gauges and the corresponding MDT value is shown with crosses.

Figure 11 displays the MDT solutions along the Northeast Atlantic coastline, from Norway to the southern tip of Spain following the labelling. The blue and red curves represent the coastal MDTs based on a combination of satellite and tide gauge MDT, and satellite only, respectively. For comparison, we have included the comparison to a typical geodetic MDT. Here we used a MDT derived in the same way as the DTU13MDT (Knudsen et al., 2011) namely based on the altimetric MDT values of MSS-geoid filtered with a Gaussian filter with radius 0.7 degrees and with no land values (shown in green). There is no doubt, that the filtering improves the stability of the MDT solution and that the various filtered solutions generally agree. In some places there are deviations up to approximately 10 cm e.g. index 350-450, 500-650, and 800-900, which corresponds to the coastline along the inner Danish waters, the North Sea, and the Bay of Biscay. However, it is difficult to quantify the quality of the GOCE++ approach visually from this plot.

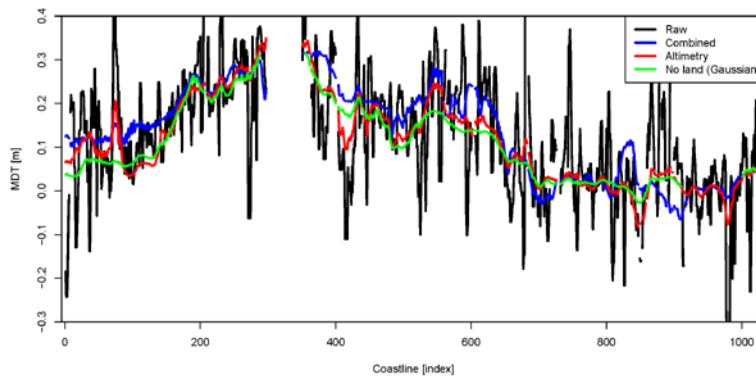


Figure 11: MDT solutions along the Northeast Atlantic.

To evaluate the uncertainty of the MDT solution we apply a bootstrap approach. Since the individual grid points are correlated, we divide the raw MDT data set into blocks, which we assume are uncorrelated. We then create a large number N of artificial data sets by sampling with replacement among the different blocks. For each of the bootstrap data set we derive a filtered MDT solution. In this way, we get a distribution of solutions from which we can estimate the standard deviation of the MDT.

For the validation we select the tide gauges along the Danish, German and French coastlines as the representation of tide gauges is dense here (corresponding to coastal grid points labelled between 300 and 900 in Figure 10). To have independent data for the validation we randomly divide the tide gauges into two groups; 20% for validation and 80% which are used to derive the MDT solution. We can then estimate the RMSE as a measure for the quality of the different solutions. Since the result will depend on which of the tide gauges that are used to derive the solution, we repeat the exercise 100 times. This provides a distribution of the RMSE for the different solutions.

The RMSE distributions for the solutions are shown in Figure 12. In the solutions, land values have been from either altimetry or a combination of altimetry and tide gauges both have medians of 5.8 cm. The MDT without land values filtered with a classical Gaussian filter of 0.7 degrees exhibits a median of 6.3 cm. In this investigation we notice a clear improvement if we include land-based MDT values or not. However, we only see a marginal difference if we use altimetry; tide gauge or combined altimetry and tide gauge MDT as land values.

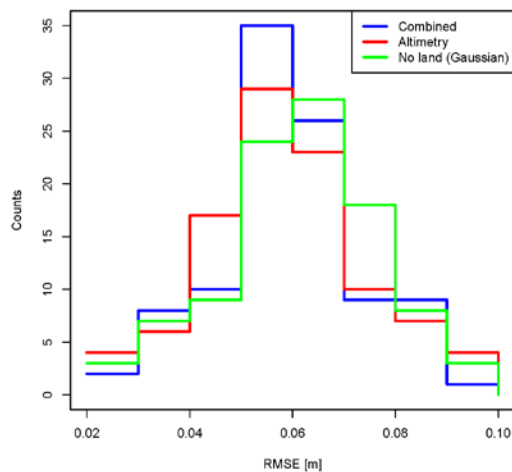


Figure 12: RMSE distributions for the MDT solutions; The MDT using combined tide gauge and altimetry as land values are shown in blue. The combined MDT using altimetry as land values is shown in red. The Gaussian filtered MDT without land values are shown in green.

5.2 Cryosat-2 validation of GOCE++ MDT

Currently SAR altimetry is becoming available with Cryosat-2 and Sentinel-3A/3B to improve the mapping of the short scales in the MSS (Huang, 2017). Hence we evaluate the potential of using SAR altimetry in future geodetic MDT models as SAR altimetry has not been used for MSS computation. In the coastal zone, satellite altimeter data processed with SAR methodology provide water height observations of higher resolution and accuracy compared to the conventional pulse-limited altimeters (Dinardo et al., 2017). Hence, we expect that they should improve the estimation of the MDT in the coastal stripe within 100 kilometres from land. We assess the CryoSat-2 altimetry products over a time interval of 6-years from October 2010 to November 2017 along the North-eastern coasts of the Atlantic Ocean to quantify this.

The SAR data are from ESA Grid Processing on Demand (G-POD) processor enables processed with two different SAR retracking methodologies, called SAMOSA-2 (Ray et al., 2014) and SAMOSA+ (Dinardo et al., 2017). The second provides improved coastal data compared to the first, differences are both in the waveforms generation (Hamming weighting window on the burst data prior to the azimuth Fast Fourier Transform (FFT), zero-padding prior to the range FFT, doubling of the extension for the radar range swath) and in the retracking methodology. We use here SAMOSA+ and its corresponding Pseudo-LRM data, also called reduced SAR altimetry (RDSAR). This latter is generated by the TUDABO processor (<http://wiki.services.eoportal.org/tiki-index.php?page=G-POD+services>) and retracked using the TU-Darmstadt Adaptive Leading Edge Sub-waveform retracker (TALES) and Spatio-Temporal Altimeter Waveform Retracking (STAR) sub-waveform retrackers (Roscher et al., 2017, Fenoglio and Buchhaupt, 2017). The quality of RDSAR and LRM data are comparable. LRM was disregarded near coast in the estimation of past MSS. The impact of the new SAMOSA+ data quality on the MDT estimation is seen from an estimate of a new MDT by averaging the 6 years of CryoSat-2 data on a 0.25 x 0.25 degree grid. We then compare it to the reference geodetic MDT output of this project. The standard deviation of differences between the reference MDT surface and the MDT surfaces constructed

from SAMOSA+, TALES and STAR open sea and coastal data are 6.7 cm for SAMOSA+, 8.5 cm for TALES and 6.9 cm for STAR respectively. The largest difference between the reference MDT and the new MDT surfaces is in the coastal zone. Figure 13 shows the difference between the reference MDT and the new MDT surfaces. The best agreement with the reference MDT (smallest differences in the German Bight) is obtained using the SAMOSA+ data (Figure 13, left) followed by the RDSAR STAR data, while with RDSAR TALES the agreement is lower.

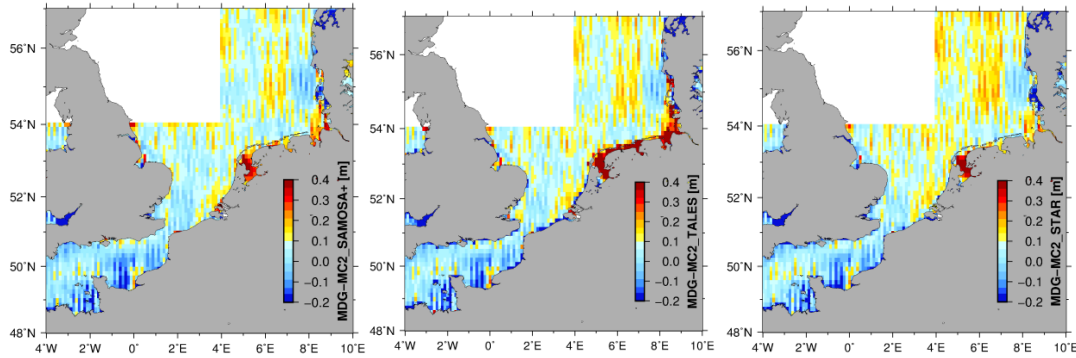


Figure. 13. Difference between the GOCE++ MDT and the MDT derived from the three SAR altimetry datasets: SAMOSA+; RDSAR (STAR and TALES) datasets.

6. Discussion, recommendations and conclusion

In the GOCE++ project we have tested a new approach to improve the MDT in the coastal zone, where the solution is based on both satellite altimetry and tide gauge data. The tide gauge MDT values are integrated into the geodetic MDT by inserting these as land values and subsequently using an iterative average filter to smooth the tide gauge MDT values into the altimetry based MDT. The approach was evaluated for the Northeast Atlantic coast where a dense representation of tide gauges is present. A validation of the coastal MDT was conducted by comparing the solution to that obtained from tide gauges. To ensure independent data for the validation only 80% of the tide gauges were included in the MDT solution. The new approach showed a small but clear improvement in terms of RMSE compared to the classical spatial Gaussian filter.

In this investigation, the land values were assigned using linear interpolation, which is not optimal when the tide gauges are unevenly distributed along the coastline or if the distance between them is large (see the heterogeneous distribution of the tide gauges on Figure 2). An improved future approach could be to consider the correlation pattern

potentially from ocean model MDTs when interpolating the land values. The GOCE++ approach to derive a coastal MDT based on altimetry and tide gauge data was implemented as an R package “coastMDT” which is freely available for further research. The package can be used with the provided test data used here or with data provided by the user.

With a “typical” misfit, presumed to be mainly due to small scale geoid errors, of 9 cm, a single tide gauge places only weak constraints on the coastal dynamic topography, especially when the non-Gaussian nature of the errors means that much larger misfits are possible. Optimal use of the tide gauges thus relies on exploiting the coherence of the MDT along the coast, together with a good quantification of errors in the tide gauge values. Preliminary analyses have shown that sea level variations at tide gauges bear very different relationships to nearby open ocean values from satellite altimetry, depending on the site considered.

Where variability at gauges agrees with open ocean altimetry, it seems safe to assume that the tide gauge mean should also reflect the ocean mean dynamic topography. Where it does not, there are three possibilities: the data quality may be poor, the local geoid error may be large, or local coastal processes may be important. Possible processes are wave setup and low salinity intrusions due to freshwater input from rivers, both of which have been suggested to contribute tens of centimeters in places. In fact, some gauges are far enough up rivers that they may reflect river flow more directly. Altimetry closer to the coast would help to distinguish between these possibilities.

Improved mapping of coastal currents and short scale geoid signals will be important to improve coastal MDT. This can be done through the integration with very high-resolution ocean model. Local analysis into the Norwegian MDT demonstrated the value of SAR altimetry in mapping local coastal topographies from CryoSat-2 (Idžanović et al, 2017). Along the Norwegian coast strong levelling ties between tide gauges permit a comparison, which is not dependent on direct GPS measurements at each gauge. The addition of measurements very close to the coast was found to help in avoiding errors due to coastal currents, seen in both the satellite data and ocean model. The remaining error, however, was found to be dominated by small-scale geoid error, which can only be addressed by local gravity measurements.

Similar investigations into Australian coastal MDT also highlighted the small-scale geoid error, at larger amplitude in this case, with sporadic mismatches of 0.1-0.2 m

(Filmer et al., 2018). In some cases, these were found to be in regions of complex coastal geometry (in one case the gauge is some distance from the coast). In this investigation, we found that addition of CryoSat-2 data had the potential to improve the coastal resolution. In this case, however, another issue was also identified: the availability of tide model data was found to be a limiting factor, which was overcome by switching from the GOT4.8 to the FES2012 tidal model. This highlights the importance of reassessing altimeter correction models as new, more-coastal data become available.

The biggest challenge for unified calculation of a coastal and global MDT is clearly the lack of small-scale geoid information for comparison with point measurements from tide gauges. This is a problem with varying geographical impact, which will ultimately only be overcome by use of local measurements; either of gravity in a region surrounding the point in question, or of geopotential at the point (see later).

A second challenge is the limited number of tide gauges with GPS ties (Woodworth et al., 2017). We have identified 302 such gauges in this project, in comparison to 1007 datum-controlled records, which overlap the satellite altimetry era (1993 onwards). The vertical velocity of the GNSS station used was known for only 141 of these 302 tide gauges (47%), the epoch of the height for only 160 (53%). Thus, for a significant number of the selected tide gauges, it was not possible to propagate the height at the chosen epoch 2005.5. Actually, 90% of the 400 coastal (closer than 20 km) ULR6 vertical velocities being in absolute value below 3.5 mm/year, it can be assumed that the maximum error of an ellipsoidal height at an unknown epoch with an unknown velocity is around 4 cm if we consider that this unknown epoch is between 1995.0 and 2017.0 (and so distant to a maximum period of 11.5 years from the chosen period 2005.5). Even without additional data, the present investigations have highlighted a number of ways in which immediate progress can be made based on observations, which are currently available or will become available in the next few years.

The accuracy of the tide gauge position can also have an impact on the MDT value through the geoid value extraction. The positions used in this study were taken from the PSMSL and SONEL databanks that strives to position the tide gauges as well as possible. To figure out the impact of possible remaining errors in the positions of the tide gauges used, a geoid slope map has been computed over a global 40 km wide coastal

strip from the EGM2008 model (Pavlis et al., 2012). The histogram on Figure 14 shows a median equal to 0.014 m/km.

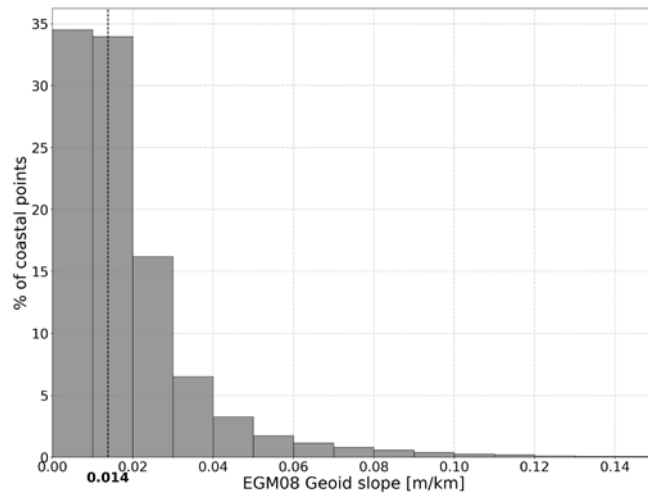


Figure 14: Histogram of the geoid slopes in coastal areas

A number of recommendations for future directions on both short-term horizon of a few years, and a longer-term decadal time scale can be found from the GOCE++ results at (<http://gocehsu.eu>). The recommendations detail our view on how to use and our strategies to improved GNSS coverage at tide including strategies to account for local vertical land movement when computing tide-gauge MDT.

The strategies for deriving an optimal MDT will play out very differently in different regions. In well-surveyed regions with many tide gauges, the tide gauge data is likely to play a significant role in constraining the coastal solutions. In poorly-surveyed regions, where tide gauge data are sparse, there will be no along-coast averaging effect, and the sporadic measurements with correspondingly large errors (given the lack of information about short length-scale variability) are unlikely to make a significant contribution. As an example, the Pacific island measurements appear to have both random and systematic errors of order 0.1 m. On the other hand, model data show that the MDT remains within 0.02 m of the coastal MDT out to distances of hundreds of kilometres from most islands meaning that the satellite data will produce the greater constraint. In order for isolated tide gauges to make a significant contribution, it is crucial that local geoid information be improved, and the improvement be quantified.

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Appendix 1.

Tide gauge locations

The locations of the 302 tide gauges used in this study, showing the numbering scheme used in the various plots within the article. The solid circles represent the first tide gauge in each colour.

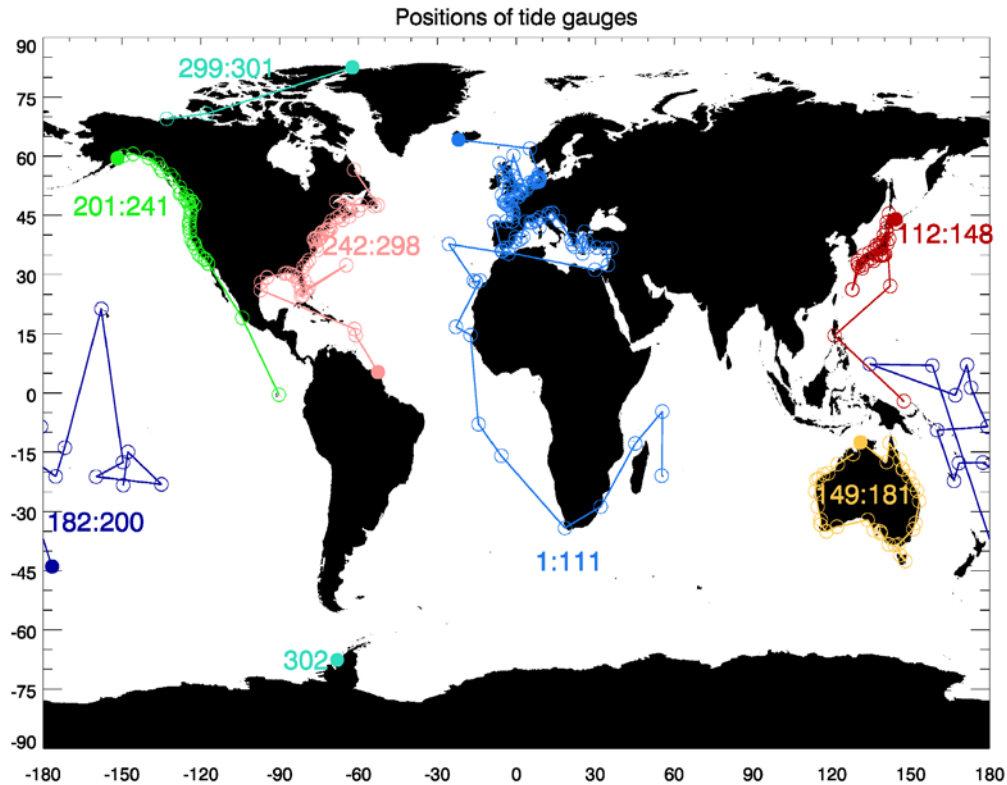


Figure A1: Positions of tide gauges used in this study, showing the numbering scheme used in plots. The solid circles represent the first tide gauge in each colour.

Appendix 2.

Software

The methodology described in the previous sections was implemented in a software package “coastMDT” written in the open source language “R”. The software is freely available from GitHub (<https://github.com/cavios/coastMDT>) including installation instructions and documentation.

The package “coastMDT” offers the user the possibility to derive a geodetic MDT for a selected region of interest. In relation to the package a collection of data sets

is also available including the MSS/MDT values at the tide gauge stations for the combined GOCO05C and EIGEN 6C4 geoids as well as the DTU 15 MSS. The gridded datasets are all given with a resolution of a $1/8 \times 1/8$ degree. A detailed description of the functions in the package is found in the CoastMDT user manual and complete example of how to use the package to derive the MDT is described in the coastMDT tutorial.